



Fondazione Policlinico Universitario A. Gemelli Università Cattolica del Sacro Cuore

#### Advanced Radiation Therapy

## Utilizzo non convenzionale dell'Imaging

## in Radioterapia

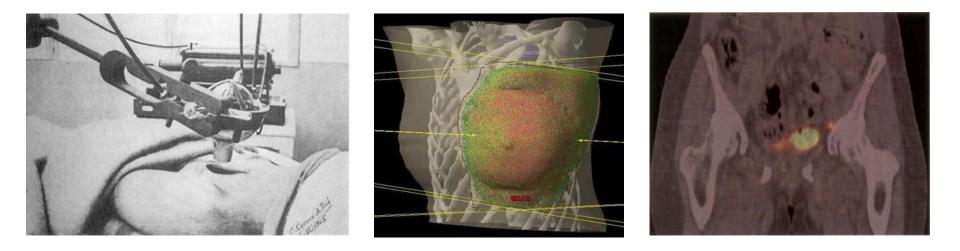
Chieti, 24 febbraio 2017



Dr. Luca Boldrini



## Conventional



## Role of imaging in Radiation Oncology

- Diagnosis
- **Staging** : locoregional, systemic
- **Characterization** : Mp imaging, IB, hybrid approaches
- **Prognostic evaluation**
- **Conventional Radiation Oncology purposes** 
  - segmentation
  - planning
  - delivery, motion management and adaptive approach acute toxicity
- Follow up : response, relapse and late toxicity



# The use of diagnostics to tailor therapeutic approaches thus facilitating personalized medicine

Bentzen S. - Lancet Oncology 2005

## Theragnostic imaging paradigm

#### **BEFORE**



Diagnosis Staging

Target volumes & Planning

DURING





In room imaging (IGRT)

Off-line / On-line

**Theragnostic Imaging** 

(biologically adapted prescription)

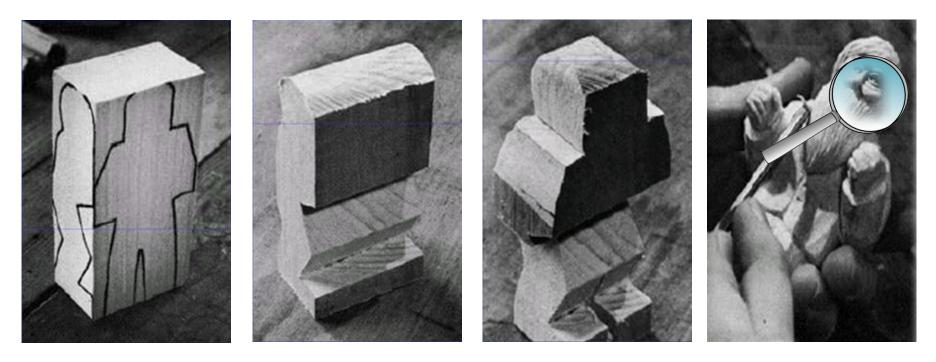
**AFTER** 



Response evaluation Tumor recurrence Late toxicity

Modified from Bentzen S. - Lancet Oncology 2005

## High technology opportunities



#### **2** D Planning

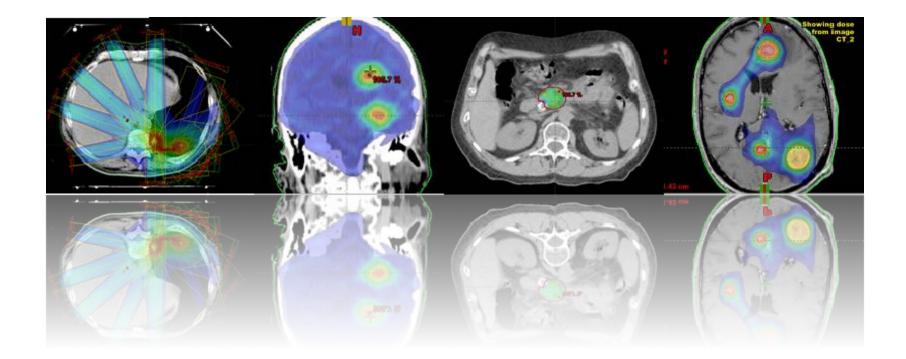
#### **3 D Planning**

#### **IMRT** Planning

### **Dose sculpting**

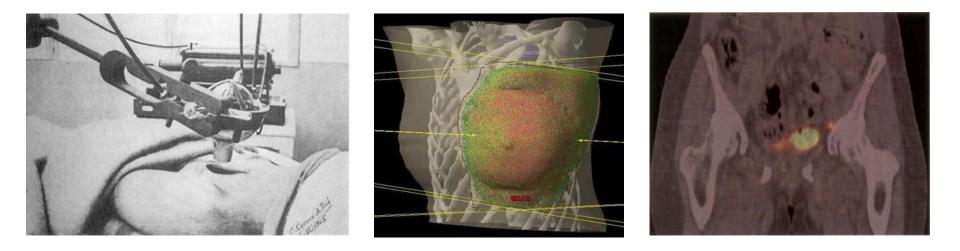
Courtesy of Verellen D. - 2013

## High technology opportunities



### Towards small (and differently visible) target volumes

## **UNconventional**



## In-room MRI: the great challenge





#### Magnetic Resonance–Guided Adaptive Radiotherapy: A Solution to the Future



Patrick Kupelian, MD,\* and Jan-Jakob Sonke, PhD<sup>+</sup>

Magnetic resonance imaging–guided adaptive radiotherapy would make available the best in anatomical and functional imaging during the course of radiation therapy. The possible methodology of magnetic resonance imaging–guided adapted radiotherapy and possible clinical applications are discussed.

Semin Radiat Oncol 24:227-232 © 2014 Elsevier Inc. All rights reserved.

## MRI-RT: an hybrid answer

Main objective: higher dose delivery to targets and toxicity reduction with organs at risk sparing through image optimization

Metabolic and functional imaging: new targets, dose painting, new toxicity paradigms

Adaptive therapy: intra- interfraction, movement management, autosegmentation

## MRI-RT: the great challenge

#### MRI in treatment room requires a fully integrated solution:

1. MRI – Linac
designed in UMC – Utrecht
8 MV accelerator, FFF
Modified 1.5 T Philips Ingenia MRI
Linac mounted in ring around MRI



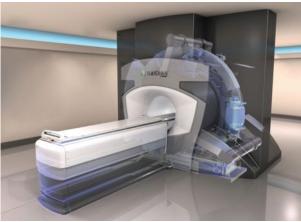


Courtesy of Uulke van der Heide

## MRI-RT: the great challenge

### MRI in treatment room requires a fully integrated solution:

2. MRI – <sup>60</sup>Co MRIdian<sup>®</sup> (ViewRay) 0.35 T MRI split magnet Real time imaging 4 frames per second 3 <sup>60</sup>Co heads (15.000 Ci each) on a ring gantry Bore size : 69.3 cm Primary collimators directly under the sources MLC : 30 leaves





## MRI–<sup>60</sup>Co RT: imaging features

Scan Name	Sequence Types	Function
Pilot Scan	GRE (3D)	Localization of anatomy and patient positioning
	TRUFI (3D)	
Planning Scan	GRE (3D)	Treatment Planning
	TRUFI (3D)	
	TFL <mark>(</mark> 3D)	
Treatment Scan	EPI (2D)	
	GRE (2D)	MRIS monitoring during treatment delivery
	TRUFI (2D)	
QA	SE (2D)	SNR, uniformity, contrast, and other QA functions

GRE: Gradient Echo - Proton density, T1, T2 - 2D GRE is 25 sec per image

TRUFI: TRUe Fast Imaging with steady state free precession – T1, T2 – 25 sec 3D planning/pilot, 0.25 sec treatment scan

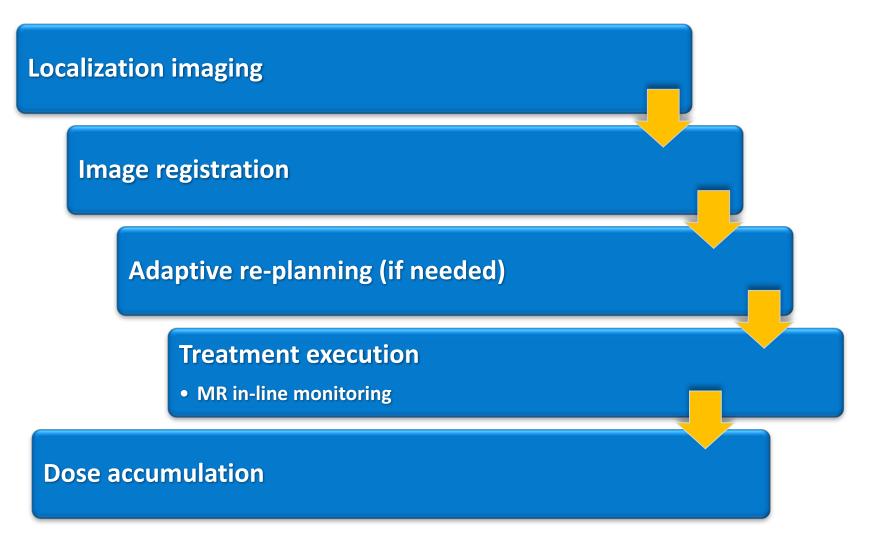
TFL: Turbo Flash – T1, mix T1/T2 – 3 min

EPI: Echo Planar Imaging – T2, mix T1/T2 – 0.25 sec per frame

SE: Spin Echo

Courtesy of VIewRay: 00016 technical manual revG

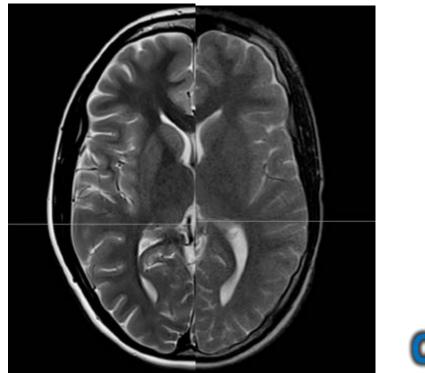
## <sup>60</sup>Co MRI-IGRT: pitfalls and questions



## <sup>60</sup>Co MRI IGRT: pitfalls and questions

- Energy of of 60Co γ-rays lower than energy LINAC X rays
  - Collimation
  - Conformality
  - Coils attenuation
  - Magnetic field perturbation
  - Patient heating
- QA of delivery-tracking process
  - Image quality and distortion
  - Autocontouring
  - Registration (elastic or rigid)
  - Eventual replanning

## <sup>60</sup>Co MRI IGRT: image quality







from http://www.healthcare.siemens.com

## <sup>60</sup>Co MRI-IGRT: pitfalls and questions

- MRI for in-room imaging opens a new era in radiation treatment workflow
- This new technology brings many expectations and multiple critical issues
- Need to multi-centric cooperation, common lexicon for MRI-RT
- Possibility to have a new tool for prognostic evaluation during the treatment execution
- Need to create a robust QA for dose accumulation algorithms
- Evaluation of impact of accumulated doses on outcome prediction

### **GAMMA team research topics**

## Gemelli Art MRI-RT team

Gamma.ITV Gamma.Similarity Gamma.Adaptive Gamma.Radiomics



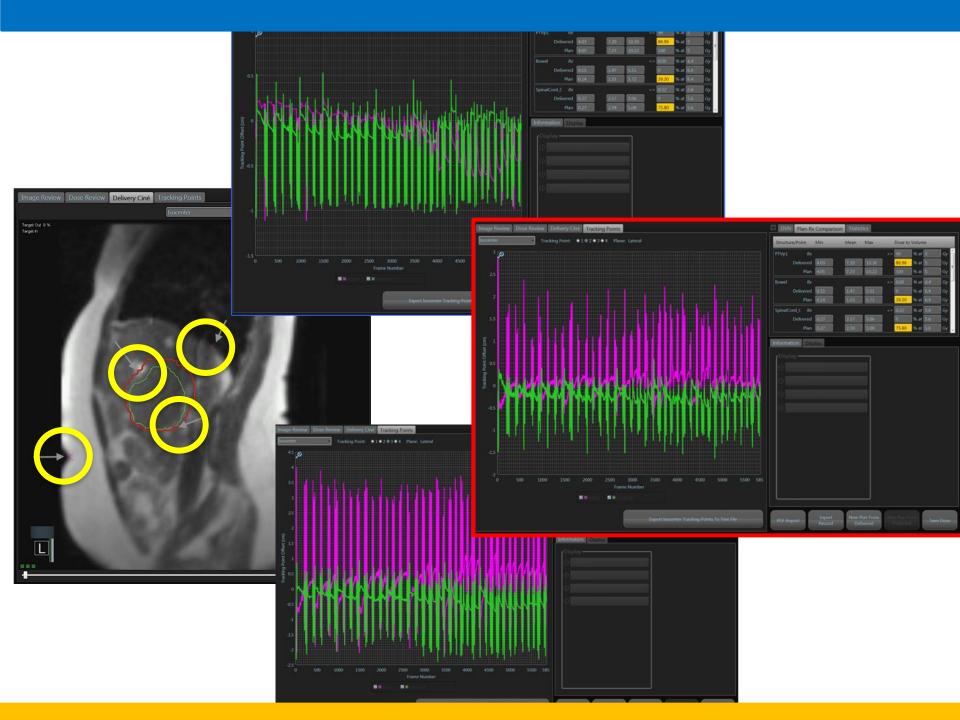


### **GAMMA.ITV: movement analysis**

Deliverable: Collaboration: Measurement of Organ Motion Specific metrics 3D/4D «Features» to MRIdian

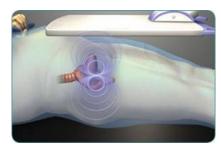




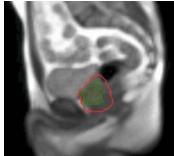


## GAMMA.ITV: movement analysis

- Organ motion (Target / OARs)
- Tailoring target volumes
- Comparative evaluation with different technologies
- Features definition for disease and patients candidate to MRIdian
- Specific metrics 3D/4D



Calipso - VARIAN



**MRIdian - VIEWRAY** 

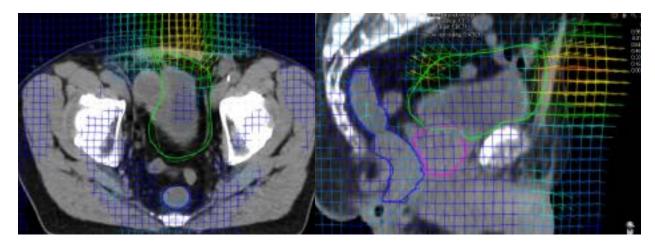
### **GAMMA.SIMILARITY**

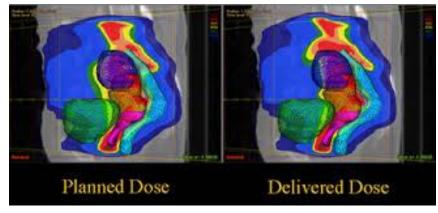
Deliverable:Acute toxicity comparisonCollaboration:Case Control Repository

- Specific metrics 3D/4D (geometrical/dosimetrical)
- Selection of case control with different technologies based on features for candidate to MRIdian
- Creation of a repository database (Multicentric)
- First endpoint to evaluate: acute toxicity comparison

### **GAMMA.ADAPTIVE: Adaptive**

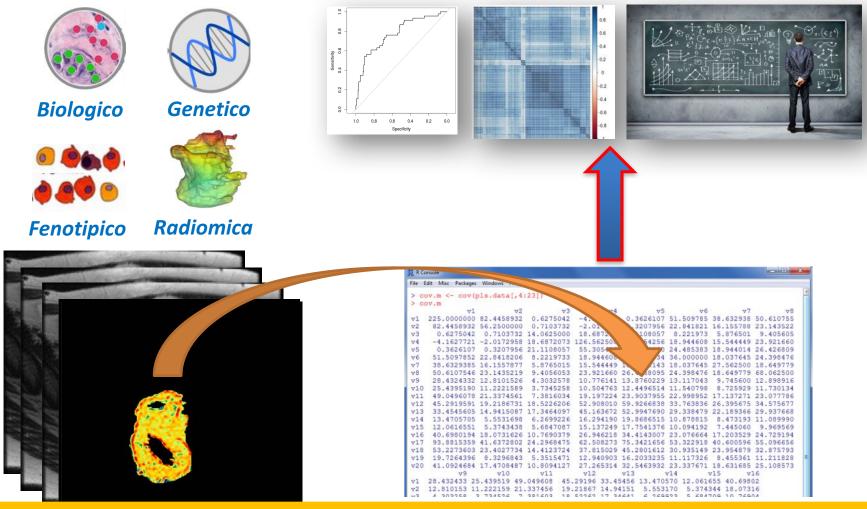
### **Deliverable: Evaluation of dose accumulation Collaboration: Dose Escalation/Reirradiation**



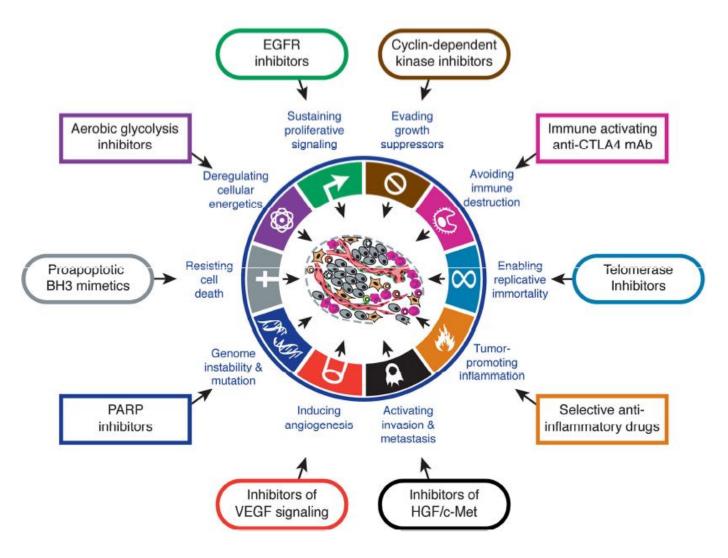


### **GAMMA.RADIOMICS**

### **Deliverable:** Radiomic analisys **Collaboration:** MRI DICOM Sharing

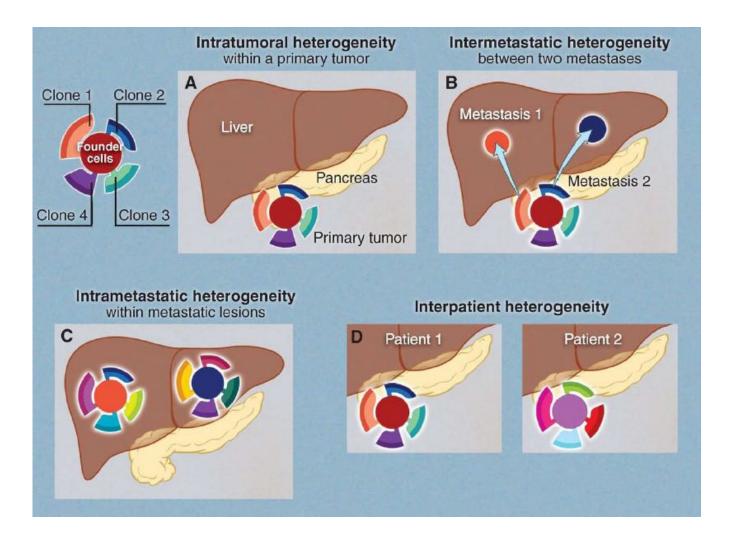


## Much more than vessels and cells...

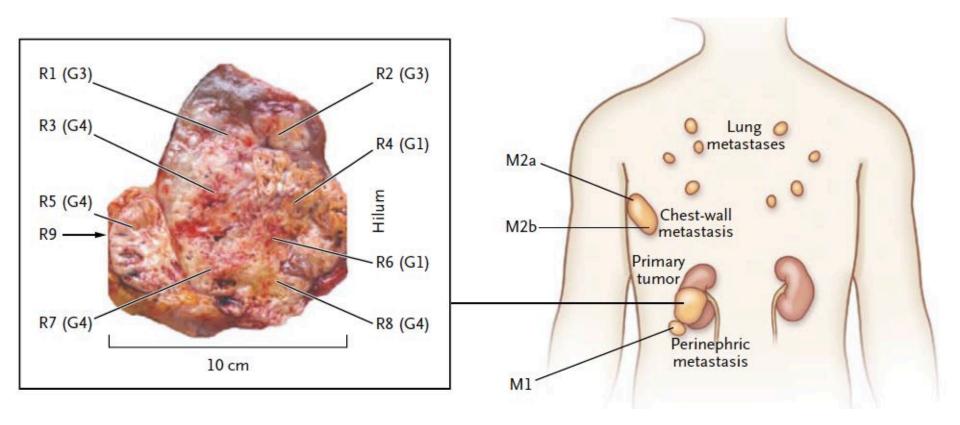


Hanahan D. and Weinberg RA. - Cell - 2011

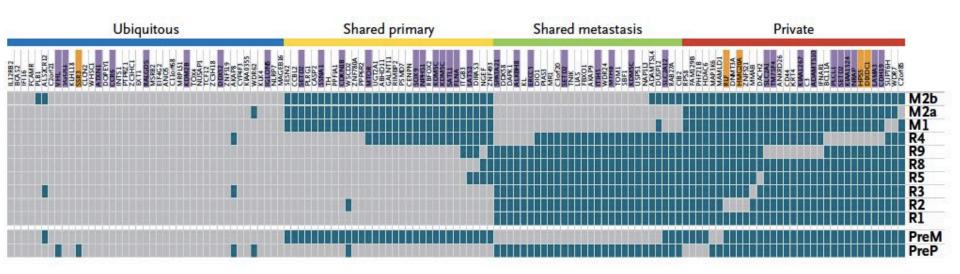
### **Tumor heterogeneity**



### **Tumor heterogeneity**



Gerlinger M. et al - NEJM - 2012



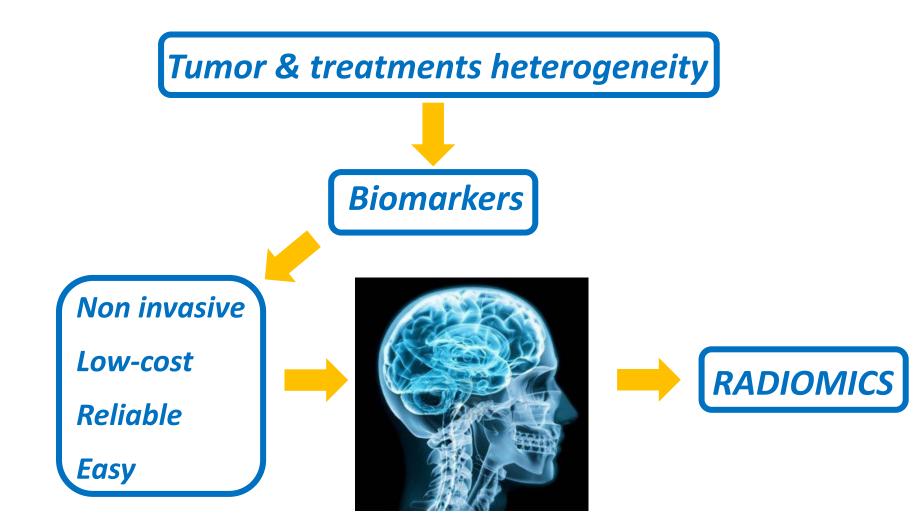
Ubiquitous	Shared	Private
CCT3 RE G3G LLCT TTN MML MML CLEC3B MML REAL REC14 REC	SETIO2 VPSI3A VPSI3A NVBPCI MVBPCI MVBPCI MVBPCI MVBPCI MVBPCI TIN TIN TIN TICIS MANI349 DCUEIA DCUEIA DCUEIA DCUEIA DCUEIA DCUEIA DCUEIA DCUEIA DCUEIA DCUEIA DCUEIA	MERTY MILLA MILLA MILLA MILLA SENIA SENIA SENIA SENIA SENIS CONDUCTION MILLA MORA MILLA MI
		R7
		R9
		R6
		R3

Gerlinger M. et al - NEJM - 2012

## **Tumor heterogeneity**



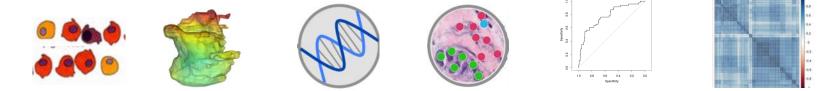
### **Tumor heterogeneity management**



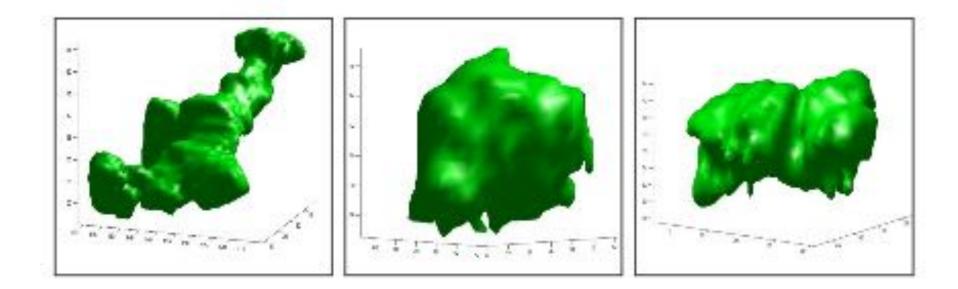
## **Radiomics**

Radiomics is the process of extraction of quantitative features from standard radiological imaging for clinical decision making tool.

Texture Analysis, Histogram Analysis and Morphometric Analysis represent the three main approaches for features extraction. Dedicated software needed.

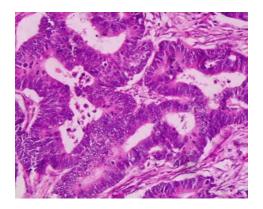


### **Radiomics: features extraction**

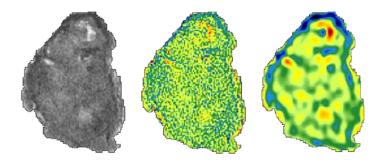


Lambin P. et al – Eur J Cancer - 2012

#### **Histological evaluation**



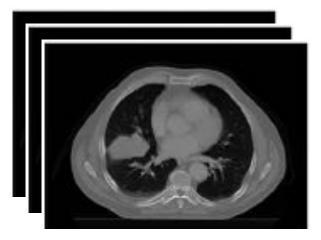
#### **Radiomics evaluation**



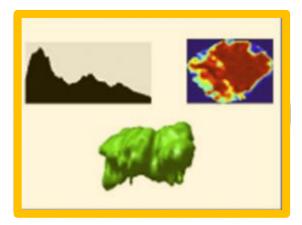
- Invasive
- Difficult to repeat
- Tumour sample not always are representative of the whole volume (tumor heterogeneity)
- Expensive

- Not invasive
- Repeatable
- Analyzes entire tumor volume
- Uses diagnostic exams already available
- Cheap

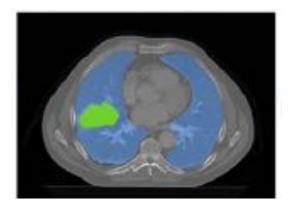
### **Radiomics analysis**



**Image collection** 



**Features extraction** 



#### Segmentation



**Analysis and modeling** 



## **Radiomics: features extraction**

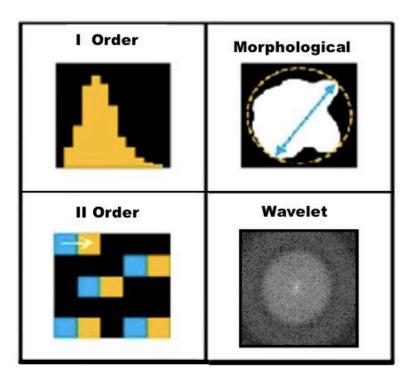
#### Features categories:

#### STATISTICAL

Mean, Median, Skewness, Entropy, Kurtosys

**TEXTURE ANALYSIS** 

Autocorrelation, Cluster Prominence, Cluster Shade





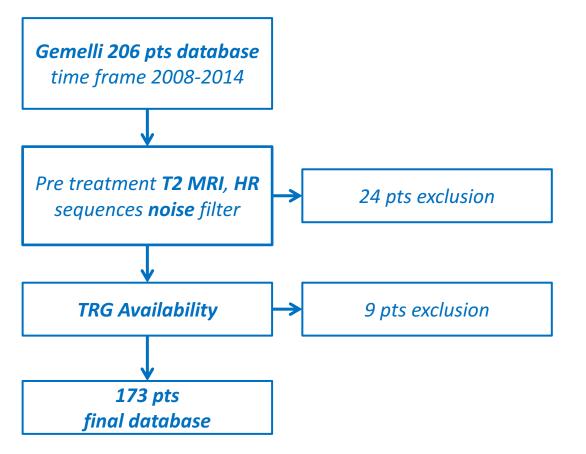


#### SHAPE

Area, Volume, Compactness, Sphericity

WAVELET

Low frequencies, high frequencies



#### **1. ROI extraction**

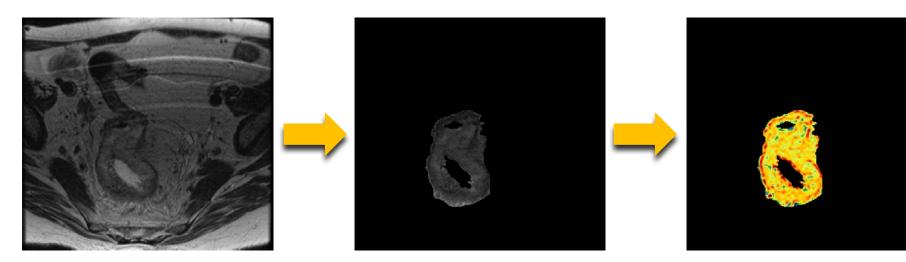
2. Pre-processing: - LoG filter application

#### 3. Data analysis (Moddicom): - Model construction

- Model validation

**ROI** extraction

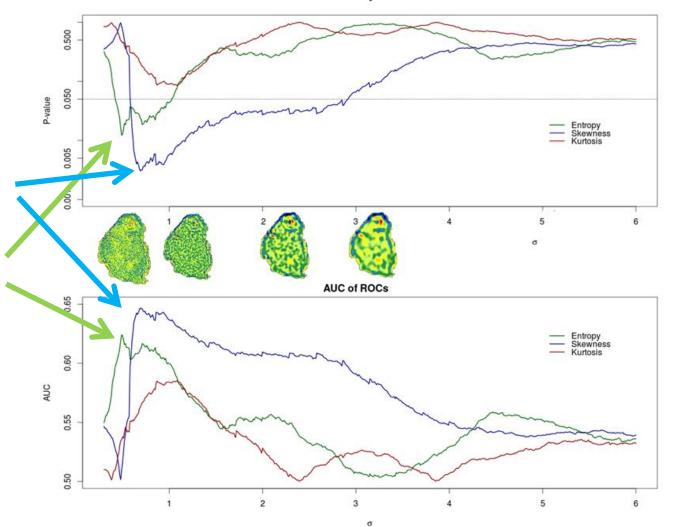
#### MRI



#### Kurtosis, Skewness, Entropy



**Filter application** 



Mann-Whitney test P-Value



The following variables were evaluated with multivariate logistic analysis for 173 rectal cancer patients

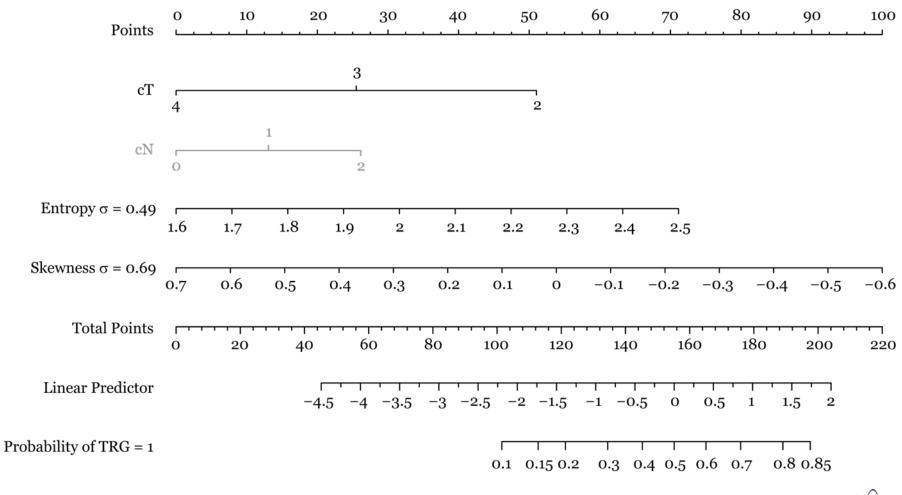
сТ	
cN	
GTV Volume	
GTV Surface	
Equivalent Sphere	Volume / GTV Surface
<b>Entropy</b> <i>s</i> = 0.49	
<i>Skewness s = 0.69</i>	
Final model:	
Coefficients:	
	Estimate Std. Error z value Pr(> z )
(Intercent)	

(Intercept)	-5.1466 3.9229 -1.312 0.18954	
сТ	-1.0442 0.3584 -2.913 <b>0.00358 **</b>	
cN	0.5350 0.3412 1.568 0.11689	
Entropy Sigma	0.49 3.2354 1.6420 1.970 0.04880 *	;
Skewness Sigma	0.69 -3.1480 1.1601 -2.714 0.00666 *	*

Datamining Radiomics KBO Labs Knowledge Based Oncology Labs

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

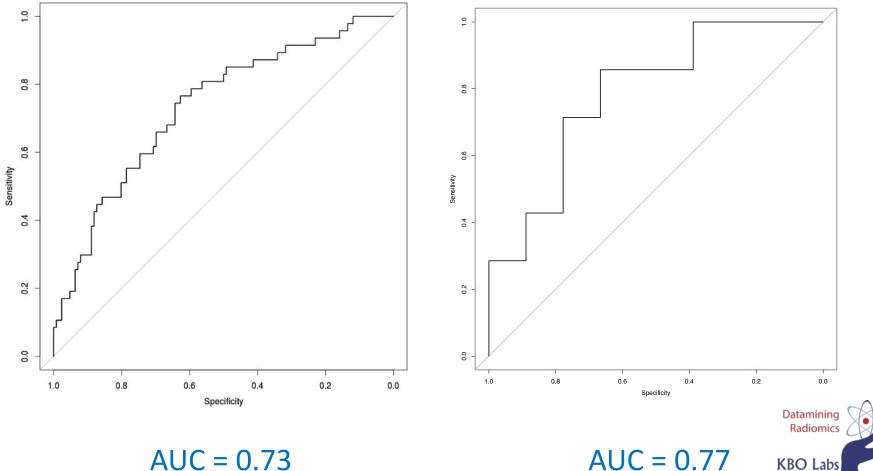
### From Radiomics to nomograms





### Internal validation 5000 bootstrap resampling TRIPOD 1b

External validation 25 cases MAASTRO TRIPOD 3



KBO Labs Knowledge Based Oncology Labs

### **Internal validation**

**External validation** 

### **KBO**

**173** Patients 47/173 pCR (28%)

**T2-w** 

Slice thickness 3 mm RM GE Signa Exite @1.5 T RM Achieva @ 1.5 T

### MAASTRO

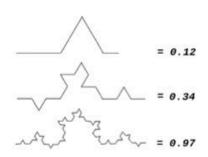
25 Patients

- 7/25 (26%)
- **T2-w**

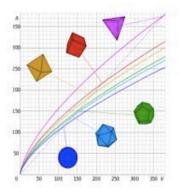
Slice thickness 3 mm



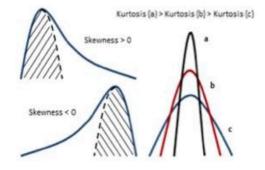
### **KBO Radiomics: features extraction**



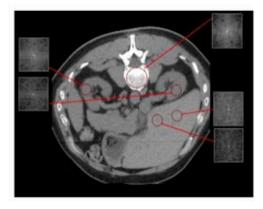
#### **Fractal dimension**



Voume/surface ratio



Grey-level Histogram Measures, (i.e. Kurtosis, Skewness, ...)

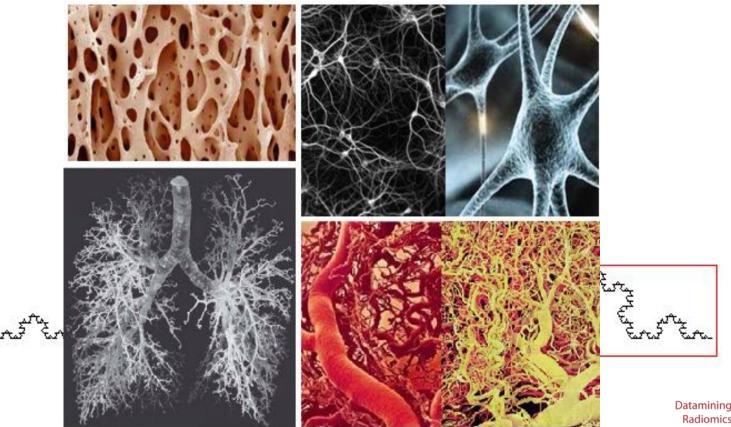


Local Granularity



### **Fractals**

### **Object that presents the same weave on different scales scale invariance**



Ponteconi et al, 2016; Cross et al, 1998; Waliszewski P, 2016



### **Fractal dimension**

The parameter that characterizes a fractal is the fractal dimension

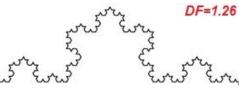
Measure of object's complexity

$$\mathsf{S}(x) = x^{DF}$$

1 < DF < 2

Low FD

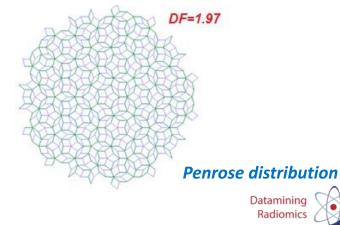
#### Pronged system



Koch curve

High FD

#### **Compact system**

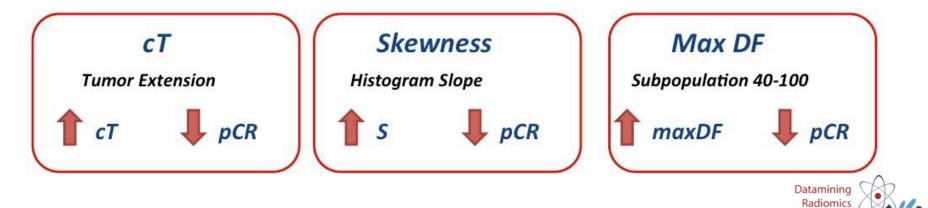


Mandelbrot B. The fractal geometry of nature. 1982 Knowledge Based Oncology Labs

## **KBO Radiomics: features extraction**

#### Introduction of Fractal Dimension

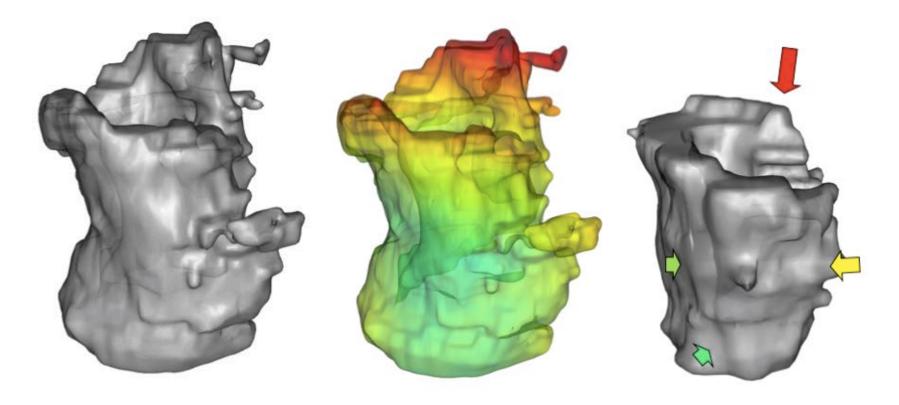
x <sub>i</sub>	α	σ (α <sub>i</sub> )	p-value
Intercept	11.366	6.119	0.063
cT	-0.997	0.377	0.007
cN	0.619	0.359	0.084
Skewness (σ=0.48)	-3.601	1.361	0.008
Entropia (σ=0.34)	2.948	1.719	0.086
Max DF (40-100)	-9.862	3.228	0.002



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### **Personalization by Radiomics**



GTV Pre

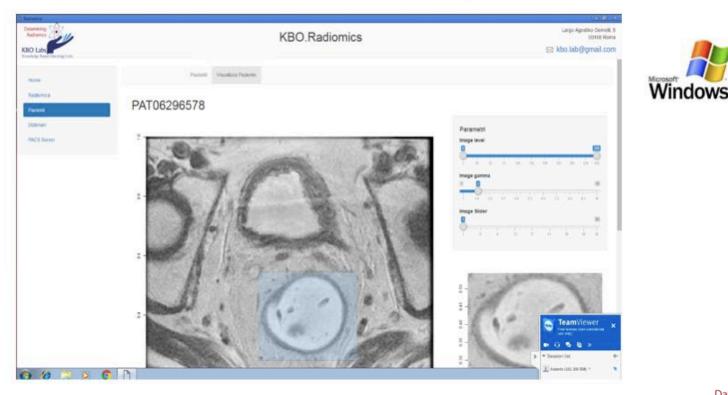
Distance heatmap

GTV Post



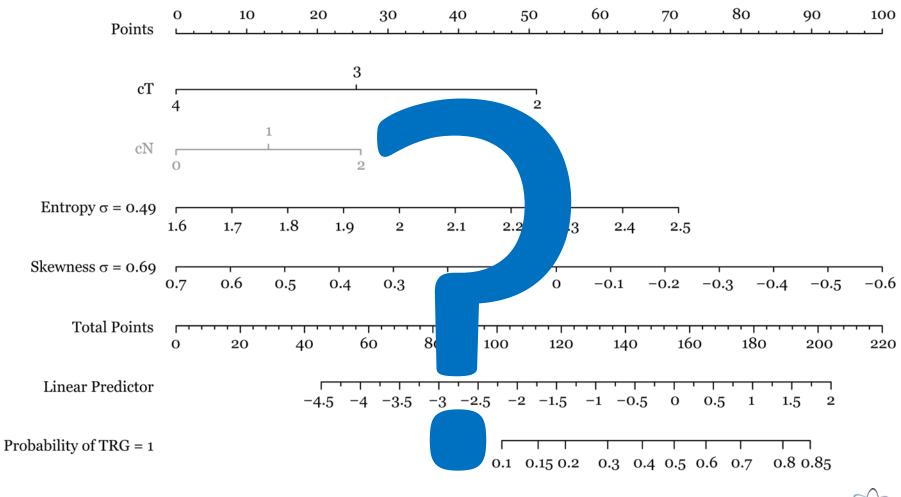
## **KBO Radiomics: features extraction**

**KBO.Radiomics** (personal edition) is a free and open source software tool, "easy to use", to facilitate the firsts steps in the world of Radiomics analysis.





### From Radiomics to nomograms



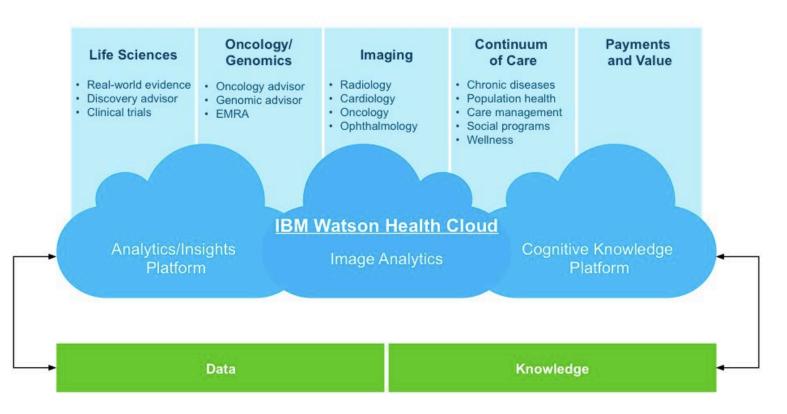


### Data sharing

Data from different sources and contexts could highly improve our knowledge



Lambin P. et al - Eur J Cancer - 2012 Valentini V. et al - J Clin Oncol - 2011 Five pillars enabled through a platform that has data, knowledge, analytics and industry specific solutions supported on a secure cloud.





#### What we would **need** to share

standards

What we are What we are

willing to share

What we are **able** 



### Data sharing

### Which barriers?

#### Table 1 Evidence for barriers to sharing of routinely collected public health data

Category	Barrier	Peer-reviewed		Non peer-reviewed
		Empirical data	Non-empirical*	
Technical	1. Data not collected	[6,21,24,31]	[2,4,7,18,22,14,26-28,30]	[3,23,25]
	2. Data not preserved		[33]	[3,32,34,35]
	3. Data not found		[45]	[3,34]
	4. Language barrier			[36]
	5. Restrictive data format		[40]	[3,34,36-39,41]
	6. Technical solutions not available		[42]	[37]
	7. Lack of metadata and standards	[21,24,43]	[40,44,45]	[1,35-37,39,41,46]
Motivational	8. No incentives		[27,45,49]	[35]
	9. Opportunity cost	[51,52]	[13,33,50,53]	[35]
	10. Possible criticism		[33]	[32]
	11. Disagreement on data use	[21]	[49]	
Economic	12. Possible economic damage		[7,26,27,30]	[55]
	13. Lack of resources	[56,21]	[13,27,28,30,42,53,57]	[3,23,34-36,39,37]
Political	14. Lack of trust	[19,59,60]	[33,61]	[34-37]
	15. Restrictive policies		[30]	
	16. Lack of guidelines		[45,62,65]	[37,41,63,64]
Legal	17. Ownership and copyright		[62,65,66,69]	[37,63,64,67]
	18. Protection of privacy	[12,19,59,73,75]	[44,57,62,66,72,74]	[36,37,64,67,68,70,71
Ethical	19. Lack of proportionality			[76]
	20. Lack of reciprocity	[51,52]	[50,77,78]	
Number of unique	documents (% of total)	14 (21.5%)	30 (46.2%)	21 (32.3%)



Willem et al BMC Public Health, 2014

## Data sharing

### **Benefits**

- transparency and cooperation
- reproducibility of research
- cost-efficiency
- preventing redundancies
- acceleration of discovery and innovation
- making more efficient and effective public health programs



# Imaging and Interventional Radiology for Radiation Oncology

Editors: Regina G.H. Beets-Tan, Wim Oyen, Vincenzo Valentini

Part I : Imaging in Oncology: from diagnosis to outcomes Part II : From simulation to delivery guided by imaging: technical aspects

Part III : Imaging for tumor staging and volume definition

Part IV : Response evaluation and Follow up by Imaging



# Looking to (a near) future

- New segmentation and planning techniques (e.g. imaging biomarkers)
- New paradigms of IGRT and adaptive real time RT
- New hybrid techniques and machines
- New prognostic stratification systems and clinical decision tools
- New radiomics perspectives and clinical integration

## **KBO Acknowledgments**

- Coordinators
  - V. Valentini, A. Damiani
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A. Dekker, J. Van Soest, P. Lambin

